# A note on the multiple unicast capacity of directed acyclic networks

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Abstract—We consider the multiple unicast problem under network coding over directed acyclic networks with unit capacity edges. There is a set of n source-terminal  $(s_i-t_i)$  pairs that wish to communicate at unit rate over this network. The connectivity between the  $s_i-t_i$  pairs is quantified by means of a connectivity level vector,  $[k_1 \ k_2 \dots k_n]$  such that there exist  $k_i$  edge-disjoint paths between  $s_i$  and  $t_i$ . Our main aim is to characterize the feasibility of achieving this for different values of n and  $[k_1 \dots k_n]$ . For 3 unicast connections (n=3), we characterize several achievable and unachievable values of the connectivity 3-tuple. In addition, in this work, we have found certain network topologies, and capacity characterizations that are useful in understanding the case of general n.

## I. INTRODUCTION

Network coding has emerged as an interesting alternative to routing in the next generation of networks. In particular, it is well-known that the network coding is a provably capacity achieving strategy for network multicast. The work of [1] provides a nice algebraic framework for reasoning about network coding, and significantly simplifies the proofs of [2], and suggests network code design schemes. However, general network connections, such as multiple unicasts are more difficult to understand under network coding. In a multiple unicast connection, there are several source terminal pairs; each source wishes to communicate to its corresponding terminal. The goal is to find a characterization of the network resources required to support this connection using network coding.

The multiple unicast problem has been examined for both directed acyclic networks [3][4][5] and undirected networks [6] in previous work. The work of [7], provides an information theoretic characterization for directed acyclic networks. However, in practice, evaluating these bounds becomes computationally infeasible even for small networks because of the large number of inequalities that are involved. Moreover, these approaches do not suggest any constructive code design approaches. The work of [4], considers the multiple unicast problem in the case of two source-terminal pairs, while the work of [3] attempts to address it by packing butterfly networks within the original graph. Das et al. [8] consider the multiple unicast problem with an interference alignment approach. For undirected networks, there is open conjecture as to whether there is any advantage to using network coding as compared to routing ([6]). Multiple unicast in the presence of

This research was supported in part by NSF grant CCF-1018148.

link faults and errors, under certain restricted (though realistic) network topologies has been studied in [9][10].

In this work our aim is to better understand the combinatorial aspects of the multiple unicast problem over directed acyclic networks. We consider a network G, with unit capacity edges and source-terminal pairs,  $s_i - t_i$ ,  $i = 1, \ldots, n$ , such that the maximum flow from  $s_i$  to  $t_i$  is  $k_i$ . Each source contains a unit-entropy message that needs to be communicated to the corresponding terminal. Our objective is to determine whether there exist feasible network codes that can satisfy the demands of the terminals. This is motivated by a need to find characterizations that can be determined in a computationally efficient manner.

## A. Main Contributions

- For the case of three unicast sessions (n=3), we identify all feasible and infeasible connectivity levels  $[k_1 \ k_2 \ k_3]$ . For the feasible cases, we provide efficient linear network code assignments. For the infeasible cases, we provide counter-examples, i.e., instances of graphs where the multiple unicast cannot be supported under any (potentially nonlinear) network coding scheme.
- We identify certain feasible/infeasible instances with two unicast sessions, where the message entropies are different. These are used to arrive at conclusions for the problem in the case of higher n (> 3).

This paper is organized as follows. In section II, we introduce several concepts that will be used throughout the paper. We also describe the precise problem formulation. Section III identifies the feasible routing connectivity levels. We discuss the network coding case in Section IV. Counter examples are given for infeasible connectivity levels. A feasible connectivity level with vector network coding solution is also provided. Section V concludes the paper.

# II. PRELIMINARIES

We represent the network as a directed acyclic graph G=(V,E). Each edge  $e\in E$  has unit capacity and can transmit one symbol from a finite field of size q per unit time (we are free to choose q large enough). If a given edge has higher capacity, it can be treated as multiple unit capacity edges. A directed edge e between nodes i and j is represented as (i,j), so that head(e)=j and tail(e)=i. A path between two nodes i and j is a sequence of edges  $\{e_1,e_2,\ldots,e_k\}$  such that

 $tail(e_1) = i, head(e_k) = j \text{ and } head(e_i) = tail(e_{i+1}), i = j$  $1, \ldots, k-1$ . The network contains a set of n source nodes  $s_i$  and n terminal nodes  $t_i, i = 1, \dots n$ . Each source node  $s_i$ observes a discrete integer-entropy source, that needs to be communicated to terminal  $t_i$ . Without loss of generality, we assume that the source (terminal) nodes do not have incoming (outgoing) edges. If this is not the case one can always introduce an artificial source (terminal) node connected to the original source (terminal) node by an edge of sufficiently large capacity that has no incoming (outgoing) edges.

We now discuss the network coding model under consideration in this paper. For the sake of simplicity, suppose that each source has unit-entropy, denoted by  $X_i$ . In scalar linear network coding, the signal on an edge (i, j), is a linear combination of the signals on the incoming edges on i or the source signals at i (if i is a source). We shall only be concerned with networks that are directed acyclic and can therefore be treated as delay-free networks [1]. Let  $Y_{e_i}$  (such that  $tail(e_i) = k$  and  $head(e_i) = l$ ) denote the signal on edge  $e_i \in E$ . Then, we have

$$Y_{e_i} = \sum_{\{e_j \mid head(e_j) = k\}} f_{j,i} Y_{e_j} \text{ if } k \in V \setminus \{s_1, \dots, s_n\}, \text{ and }$$

$$\begin{split} Y_{e_i} &= \sum_{\{e_j \mid head(e_j) = k\}} f_{j,i} Y_{e_j} \text{ if } k \in V \backslash \{s_1, \dots, s_n\}, \text{ and} \\ Y_{e_i} &= \sum_{j=1}^n a_{j,i} X_j \quad \text{where } a_{j,i} = 0 \text{ if } X_j \text{ is not observed at } k. \end{split}$$

The coefficients  $a_{j,i}$  and  $f_{j,i}$  are from the operational field. Note that since the graph is directed acyclic, it is equivalently possible to express  $Y_{e_i}$  for an edge  $e_i$  in terms of the sources  $X_j$ 's. If  $Y_{e_i} = \sum_{k=1}^n \beta_{e_i,k} X_k$  then we say that the global coding vector of edge  $e_i$  is  $\beta_{e_i} = [\beta_{e_i,1} \cdots \beta_{e_i,n}]$ . We shall also occasionally use the term coding vector instead of global coding vector in this paper. We say that a node i (or edge  $e_i$ ) is downstream of another node j (or edge  $e_i$ ) if there exists a path from j (or  $e_i$ ) to i (or  $e_i$ ).

Vector linear network coding is a generalization of the scalar case, where we code across the source symbols in time, and the intermediate nodes can implement more powerful operations. Formally, suppose that the network is used over T time units. We treat this case as follows. Source node  $s_i$  now observes a vector source  $[X_i^{(1)} \ldots X_i^{(T)}]$ . Each edge in the original graph is replaced by T parallel edges. In this graph, suppose that a node j has a set of  $\beta_{inc}$  incoming edges over which it receives a certain number of symbols, and  $\beta_{out}$  outgoing edges. Under vector network coding, j chooses a matrix of dimension  $\beta_{out} \times \beta_{inc}$ . Each row of this matrix corresponds to the local coding vector of an outgoing edge from j.

Note that the general multiple unicast problem, where edges have different capacities and the sources have different entropies can be cast in the above framework by splitting higher capacity edges into parallel unit capacity edges, a higher entropy source into multiple, collocated unit-entropy sources; and the corresponding terminal node into multiple, collocated terminal nodes.

An instance of the multiple unicast problem is specified by the graph G and the source terminal pairs  $s_i - t_i$ , i = 1, ..., n,

and is denoted  $< G, \{s_i - t_i\}_1^n, \{R_i\}_1^n >$ , where the rates  $R_i$ denote the entropy of the  $i^{th}$  source. For convenience, if all the sources are unit entropy, we will refer to the instance by just  $\{G, \{s_i - t_i\}_{1}^n \}$ , where the  $s_i - t_i$  connections will occasionally be referred to as sessions that we need to support.

The instance is said to have a scalar linear network coding solution if there exist a set of linear encoding coefficients for each node in V such that each terminal  $t_i$  can recover  $X_i$ using the received symbols at its input edges. Likewise, it is said to have a vector linear network coding solution with vector length T if the network employs vector linear network codes and each terminal  $t_i$  can recover  $[X_i^{(1)} \dots X_i^{(T)}]$ .

We will also be interested in examining the existence of a routing solution, wherever possible. In a routing solution, each edge carries a copy of one of the sources, i.e., each coding vector is such that at most one entry takes the value 1, all others are 0. Scalar (vector) routing solutions can be defined in a manner similar to scalar (vector) network codes. We now define some quantities that shall be used throughout the paper.

Definition 2.1: Connectivity level. The connectivity level for source-terminal pair  $s_i - t_i$  is said to be n if the maximum flow between  $s_i$  and  $t_i$  in G is n. The connectivity level of the set of connections  $s_1 - t_1, \ldots, s_n - t_n$  is the vector  $[\max\text{-flow}(s_1-t_1) \max\text{-flow}(s_2-t_2) \dots \max\text{-flow}(s_n-t_n)].$ 

In this work our aim is to characterize the feasibility of the multiple unicast problem based on the connectivity level of the  $s_i - t_i$  pairs. The questions that we seek to answer are of the following form.

Suppose that the connectivity level is  $[k_1 \ k_2 \ \dots \ k_n]$ . Does any instance always have a linear (scalar or vector) network coding solution? If not, is it possible to demonstrate a counterexample, i.e, an instance of a graph G and  $s_i - t_i$ 's such that recovering  $X_i$  at  $t_i$  for all i is impossible under linear (or nonlinear) strategies?

In this paper, our achievability results will be constructive and based on linear network coding, whereas the counterexamples will hold under all possible strategies.

# III. MULTIPLE UNICAST UNDER ROUTING

We begin by providing a simple condition that guarantees the existence of a routing solution.

Theorem 3.1: Consider a multiple unicast instance with n  $s_i - t_i$  pairs such that the connectivity level is  $[n \ n \ \dots \ n]$ . There exists a vector routing solution for this instance.

*Proof:* Under vector routing over n time units, source  $s_i$ observes  $[X_i^{(1)} \dots X_i^{(n)}]$  symbols. Each edge e in the original graph is replaced by n parallel edges,  $e^1, e^2, \ldots, e^n$ . Let  $G_{\alpha}$ represent the subgraph of this graph consisting of edges with superscript  $\alpha$ . It is evident that max-flow $(s_{\alpha} - t_{\alpha}) = n$  over  $G_{\alpha}$ . Thus, we transmit  $X_{\alpha}^{(1)}, \ldots, X_{\alpha}^{(n)}$  over  $G_{\alpha}$  using routing, for all  $\alpha = 1, ..., n$ . It is clear that this strategy satisfies the demands of all the terminals.

Note that in general, a network with the above connectivity level may not be able to support a scalar routing solution, an instance is shown in Figure 1. However, a scalar network coding solution exists for this example.

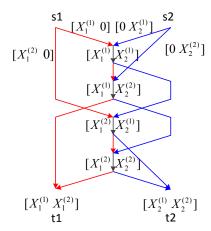


Fig. 1. A network with connectivity levels  $[2\ 2]$  and rate  $\{1,1\}$ . There is a vector routing solution as shown in the figure. There is no scalar routing solution.

#### IV. NETWORK CODING FOR THREE UNICAST SESSIONS

In the case of three unicast sessions, it is clear based on the results of Section III that if the connectivity level is  $[3\ 3\ 3]$ , then a vector routing solution always exists. In this section we provide a full characterization of the feasibility/infeasibility of supporting three unicast sessions for a connectivity level of  $[k_1\ k_2\ k_3]$ , where  $1\le k_i,\le 3, i=1,\ldots,3$ . For the feasible cases we will demonstrate appropriate linear network code assignments. On the other hand, for the infeasible cases we will present counter-examples where it is not possible to satisfy the terminal's demands under any coding strategy.

## A. Infeasible Instances

We begin by demonstrating certain instances that can be ruled out by using cutset bounds.

Lemma 4.1: There exist multiple unicast instances with three unicast sessions such that the connectivity levels  $[2\ 2\ 2]$  and  $[1\ 1\ 3]$  are infeasible.

*Proof:* A network with connectivity levels  $[2\ 2\ 2]$  is shown in Figure 2(a). Consider the cut specified by the set of nodes  $\{s_1, s_2, s_3, v_1, v_2\}$  that has a capacity value of 2. The rate that needs to be supported over  $\{e_1, e_2\}$  is 3. By the cut set bound, this rate cannot be achieved.

Similarly, a network with connectivity levels  $[1 \ 1 \ 3]$  is shown in Figure 2(b). Consider the cut  $\{s_1, s_2, v_1\}$ . The capacity of this cut is 1. However, the rate that needs to be supported over  $e_1$  is 2. Therefore, there does not exist a network coding solution.

While cut set bound is useful in the above cases, there exist certain connectivity levels for which a cut set bound is not tight enough. We now present such an instance in Figure 3. We show that this instance is not feasible under any code scheme (linear or nonlinear). This instance was also presented in the work of Erez and Feder [11], though they did not provide a formal proof of this fact.

Lemma 4.2: There exists a multiple unicast instance, with two sessions  $< G, \{s_1 - t_1, s_2 - t_2\}, \{2, 1\} >$  and connectivity

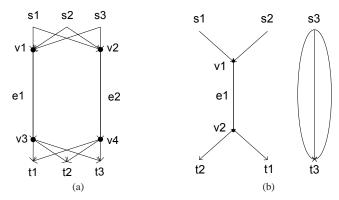


Fig. 2. (a) An example of  $[2\ 2\ 2]$  connectivity network without a network coding solution. (b) An example of  $[1\ 1\ 3]$  connectivity network without a network coding solution.

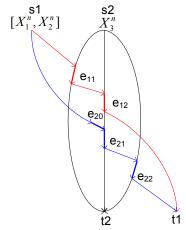


Fig. 3. An example of  $[2\ 3]$  connectivity network, rate  $\{2,1\}$  cannot be supported.

level [2 3] that is infeasible.

Proof: The graph instance is shown in Figure 3. Assume in n time units,  $s_1$  observes two independent vector sources  $[X_1^{(1)} \ldots X_1^{(n)}]$  and  $[X_2^{(1)} \ldots X_2^{(n)}]$ ,  $s_2$  observes one independent vector source  $[X_3^{(1)} \ldots X_3^{(n)}]$ . The sources are denoted as  $X_1^n$ ,  $X_2^n$  and  $X_3^n$  for simplicity. The n random variables that  $e_i$  carries are denoted as  $Y_{e_i}^n$ , or simply  $Y_i^n$ . Suppose that the alphabet of  $X_i$  is  $\mathcal{X}$ . Since the entropy rates for the three sources are the same, we can assume  $H(X_i) = \log |\mathcal{X}| = a$ . Also, since we are interested in the feasibility of the solution, we can further assume that the alphabet size of  $Y_{ij}$  is also the same as  $\mathcal{X}$ , and  $H(Y_{ij}) \leq \log |\mathcal{X}| = a$  by the capacity constraint of the edge. At terminal  $t_1$  and  $t_2$ , from  $Y_{11}^n$ ,  $Y_{12}^n$ ,  $Y_{21}^n$  and  $Y_{22}^n$ , we estimate  $X_1^n$ ,  $X_2^n$  and  $X_3^n$ . Let the estimate be  $\widehat{X}_1^n$ ,  $\widehat{X}_2^n$  and  $\widehat{X}_3^n$ . Suppose that there exist network codes and decoding function such that  $P((\widehat{X}_1^n, \widehat{X}_2^n) \neq (X_1^n, X_2^n)) \to 0$  as  $n \to \infty$ . From the Fano's inequality, we shall have

$$H(X_1^n, X_2^n | \widehat{X}_1^n, \widehat{X}_2^n) \le n\epsilon_n. \tag{1}$$

where  $n\epsilon_n = 1 + nP_e \log(|\mathcal{X}|)$ . For  $t_1$  to decode  $X_1^n$  and  $X_2^n$  asymptotically,  $\epsilon_n \to 0$  as  $P_e \to 0$ , when  $n \to \infty$ , where

$$P_e = P((\widehat{X}_1^n, \widehat{X}_2^n) \neq (X_1^n, X_2^n)).$$

Likewise, decodability at  $t_1$  implies that  $\widehat{X}_1^n, \widehat{X}_2^n$  are functions of  $Y_{12}^n$  and  $Y_{22}^n$ . Hence, we will have

$$H(X_1^n, X_2^n | Y_{12}^n, Y_{22}^n) = H(X_1^n, X_2^n | \widehat{X}_1^n, \widehat{X}_2^n, Y_{12}^n, Y_{22}^n)$$

$$\leq H(X_1^n, X_2^n | \widehat{X}_1^n, \widehat{X}_2^n) \leq n\epsilon_n.$$
(2)

Now the sequences of information coming into  $t_1$  are,

$$2an \stackrel{(a)}{\geq} H(Y_{12}^n, Y_{22}^n)$$

$$\stackrel{(b)}{=} H(Y_{12}^n, Y_{22}^n, X_1^n, X_2^n) - H(X_1^n, X_2^n | Y_{12}^n, Y_{22}^n)$$

$$\geq H(X_1^n, X_2^n) - H(X_1^n, X_2^n | Y_{12}^n, Y_{22}^n)$$

$$\stackrel{(c)}{\geq} 2an - n\epsilon_n$$

$$(3)$$

(a) is due to the capacity constraints of the edge  $e_{12}$  and  $e_{22}$ . (b) follows from the chain rule. (c) is because rate 2an should be transmitted over n time units and Equation (2).

Next, we shall have

$$H(Y_{12}^{n}, Y_{22}^{n} | X_{1}^{n}, X_{2}^{n})$$

$$\stackrel{(a)}{=} H(Y_{12}^{n}, Y_{22}^{n}, X_{1}^{n}, X_{2}^{n}) - H(X_{1}^{n}, X_{2}^{n})$$

$$\stackrel{(b)}{=} H(X_{1}^{n}, X_{2}^{n} | Y_{12}^{n}, Y_{22}^{n}) + H(Y_{12}^{n}, Y_{22}^{n}) - H(X_{1}^{n}, X_{2}^{n})$$

$$\stackrel{(c)}{\leq} n\epsilon_{n} + 2an - 2an = n\epsilon_{n}.$$

$$(4)$$

(a)(b) follows from the chain rule. (c) is from Equation (2) and Equation (3).

Analyzing the independence of  $X_1^n$ ,  $X_2^n$ , and  $X_3^n$ , we shall have

$$an = H(X_{3}^{n}|X_{1}^{n}, X_{2}^{n})$$

$$\stackrel{(a)}{=} H(X_{3}^{n}|Y_{12}^{n}, Y_{22}^{n}, X_{1}^{n}, X_{2}^{n}) + I(X_{3}^{n}; Y_{12}^{n}, Y_{22}^{n}|X_{1}^{n}, X_{2}^{n})$$

$$= H(X_{3}^{n}|Y_{12}^{n}, Y_{22}^{n}, X_{1}^{n}, X_{2}^{n}) + H(Y_{12}^{n}, Y_{22}^{n}|X_{1}^{n}, X_{2}^{n})$$

$$- H(Y_{12}^{n}, Y_{22}^{n}|X_{1}^{n}, X_{2}^{n}, X_{3}^{n})$$

$$\stackrel{(b)}{\leq} H(X_{3}^{n}|Y_{12}^{n}, Y_{22}^{n}, X_{1}^{n}, X_{2}^{n}) + n\epsilon_{n}$$

$$\stackrel{(c)}{\leq} H(X_{3}^{n}|Y_{12}^{n}, Y_{22}^{n}) + n\epsilon_{n} \stackrel{(d)}{\leq} an + n\epsilon_{n}$$

$$(5)$$

(a) is from the definition of conditional mutual information. (b) is from Equation (4) and because conditioning reduces entropy. (c) is because conditioning reduces entropy. (d) is because conditioning reduces entropy. From the above inequalities, the information on  $e_{12}$  and  $e_{22}$  cannot decode  $X_3^n$  asymptotically. Then we have the following equations,

$$an - n\epsilon_n < H(X_3^n | Y_{12}^n, Y_{22}^n) < an$$
 (6)

$$I(Y_{12}^n, Y_{22}^n; X_3^n) = H(X_3^n) - H(X_3^n | Y_{12}^n, Y_{22}^n) \le n\epsilon_n; \quad (7$$

$$H(Y_{12}^n, Y_{22}^n | X_3^n) = H(Y_{12}^n, Y_{22}^n) - I(Y_{12}^n, Y_{22}^n; X_3^n)$$

$$> 2an - 2n\epsilon_n$$
(8)

$$I(Y_{12}^n; X_3^n) = I(Y_{12}^n, Y_{22}^n; X_3^n) - I(Y_{22}^n; X_3^n | Y_{12}^n) \le n\epsilon_n;$$

$$I(Y_{22}^n; X_3^n) \le n\epsilon_n$$
(9)

The above inequalities imply that the information on  $e_{12}$  and  $e_{22}$  are asymptotically independent of  $X_3^n$ .

Because  $Y_{21}^n$  is only a function of  $Y_{12}^n$  and  $Y_{20}^n$ ,

$$H(Y_{21}^n, Y_{22}^n)$$

$$\stackrel{(a)}{=} H(X_3^n, Y_{21}^n, Y_{22}^n) - H(X_3^n | Y_{21}^n, Y_{22}^n)$$

$$\stackrel{\scriptscriptstyle{(b)}}{=} H(X_3^n, Y_{21}^n) - H(X_3^n | Y_{21}^n, Y_{22}^n)$$

$$\stackrel{(c)}{\leq} 2an - H(X_3^n | Y_{21}^n, Y_{22}^n)$$

$$\overset{^{(d)}}{\leq} 2an - H(X_3^n|Y_{21}^n,Y_{22}^n,Y_{20}^n,Y_{12}^n,X_1^n,X_2^n)$$

$$\stackrel{(e)}{=} 2an - H(X_3^n | Y_{22}^n, Y_{20}^n, Y_{12}^n, X_1^n, X_2^n)$$
(10)

$$\stackrel{(f)}{=} 2an - H(X_3^n | Y_{22}^n, X_1^n, X_2^n, Y_{12}^n)$$

$$\stackrel{(g)}{=} 2an - H(X_3^n | Y_{22}^n, Y_{12}^n) + I(X_3^n; X_1^n, X_2^n | Y_{22}^n, Y_{12}^n)$$

$$\stackrel{(h)}{=} 2an - H(X_3^n | Y_{22}^n, Y_{12}^n) + H(X_1^n, X_2^n | Y_{22}^n, Y_{12}^n) - H(X_1^n, X_2^n | Y_{22}^n, X_3^n, Y_{12}^n)$$

$$\leq 2an - H(X_3^n|Y_{22}^n, Y_{12}^n) + H(X_1^n, X_2^n|Y_{22}^n, Y_{12}^n)$$

$$\stackrel{\scriptscriptstyle (i)}{\leq} 2an - an + n\epsilon_n + n\epsilon_n = an + 2n\epsilon_n$$

(a) follows from the chain rule, (b) is because  $Y_{22}^n$  is a function of  $X_3^n$  and  $Y_{21}^n$ . (c) is because of the capacity constraints. (d) is because conditioning reduces entropy. (e) is because  $Y_{21}^n$  is a function of  $Y_{12}^n$  and  $Y_{20}^n$ . (f) is because  $Y_{20}^n$  is a function of  $X_1^n$  and  $X_2^n$ . (g)(h) follows from the mutual information definition. (i) is from Equation (2) and Equation (6). The above inequalities indicate that  $e_{21}$  and  $e_{22}$  should carry the same information asymptotically.

From the network, we know that  $Y_{12}^n$  is a function of  $Y_{11}^n$  and  $X_3^n$ . Then

$$H(Y_{11}^{n}, Y_{21}^{n}, Y_{22}^{n} | X_{3}^{n}) = H(Y_{11}^{n}, Y_{21}^{n}, Y_{22}^{n}, X_{3}^{n} | X_{3}^{n})$$

$$\geq H(Y_{12}^{n}, Y_{21}^{n}, Y_{22}^{n} | X_{3}^{n})$$

$$\geq H(Y_{22}^{n}, Y_{12}^{n} | X_{3}^{n}) \stackrel{(a)}{\geq} 2an - 2n\epsilon_{n}$$
(11)

(a) is due to Equation (8).

Finally, we shall have

$$H(X_{3}^{n}|Y_{11}^{n}, Y_{21}^{n}, Y_{22}^{n})$$

$$= H(Y_{11}^{n}, Y_{21}^{n}, Y_{22}^{n}|X_{3}^{n}) + H(X_{3}^{n}) - H(Y_{22}^{n}, Y_{21}^{n}, Y_{11}^{n})$$

$$\stackrel{(a)}{\geq} 2an - 2n\epsilon_{n} + an - H(Y_{22}^{n}, Y_{21}^{n}, Y_{11}^{n})$$

$$= 3an - 2n\epsilon_{n} - H(Y_{22}^{n}, Y_{21}^{n}) - H(Y_{11}^{n}|Y_{22}^{n}, Y_{21}^{n})$$

$$\stackrel{(b)}{\geq} 3an - 2n\epsilon_{n} - an - 2n\epsilon_{n} - H(Y_{11}^{n}|Y_{22}^{n}, Y_{21}^{n})$$

$$\stackrel{(c)}{\geq} 2an - 4n\epsilon_{n} - an = an - 4n\epsilon_{n}$$

$$(12)$$

(a) is because of Equation (11). (b) is because of Equation (10). (c) is due to the capacity constraint of  $Y_{11}^n$ .

When  $n \to \infty$ , for  $t_1$  to asymptotically decode  $X_1^n$  and  $X_2^n$ , we shall have  $\epsilon_n \to 0$ . Then  $t_2$  cannot decode  $X_3^n$  asymptotically.

Corollary 4.3: There exists a multiple unicast instance with three sessions, and connectivity level [2 3 2] that is infeasible.

*Proof:* Consider a multiple unicast instance G,  $\{s_i' - t_i'\}_1^3$ ,  $\{1, 1, 1\} >$ , where G is the graph in Figure 3. The sources  $s_1'$  and  $s_3'$  are collocated at  $s_1$  (in G), and the terminals  $t_1'$  and  $t_3'$  are collocated at  $t_1$  (in G). Likewise, the source  $s_2'$  and

terminal  $t'_2$  are located at  $s_2$  and  $t_2$  in G. The three sessions have connectivity level  $[2\ 3\ 2]$ . Based on the arguments in Lemma 4.2, there is no feasible solution for this instance.

The instance presented in Lemma 4.2, can be generalized to obtain a series of counter-examples. In particular, we have the following theorem shows an instance with two unicast sessions with connectivity level  $[n_1 \ n_2]$  that cannot support rates  $R_1 = n_1, R_2 = n_2 - n_1$ .

Theorem 4.4: For a directed acyclic graph G with two s-t pairs, if the connectivity level for  $(s_1,t_1)$  is  $n_1$ , for  $(s_2,t_2)$  is  $n_2$ ,  $1 < n_1 < n_2$ , there exist instances that cannot support  $R_1 = n_1$  and  $R_2 = n_2 - n_1$ .

*Proof:* The proof is omitted due to space limitations. ■

## B. Feasible Instances

It is evident that the infeasibility of an instance with connectivity level  $[2\ 2\ 3]$  implies that when  $1 \le k_i \le 3$ , the only possible instances that are potentially feasible are  $[1\ 3\ 3]$ , its permutations and connectivity levels that are greater than it. We now show that many of these instances are feasible using linear network codes. In this subsection, we present efficient linear network code assignment algorithms for these cases. Towards this end, we need the following definitions.

Definition 4.5: Minimality. Consider a multiple unicast instance  $G = (V, E), \{s_i - t_i\}_1^n >$ , with connectivity level  $[k_1 \ k_2 \ \dots \ k_n]$ . The graph G is said to be minimal if the removal of any edge from E strictly reduces the connectivity level. If G is minimal, we will also refer to the multiple unicast instance as minimal.

Clearly, given a non-minimal instance G=(V,E), we can always remove the non-essential edges from it, to obtain the minimal graph  $G_{\min}$ . This does not affect feasibility, since a network code for  $G_{\min}=(V,E_{\min})$  can be converted into a network code for G by simply assigning the all-zeros coding vector to the edges in  $E\backslash E_{\min}$ .

Definition 4.6: Overlap edge. An edge e is said to be an overlap edge for paths  $P_i$  and  $P_j$  in G, if  $e \in P_i \cap P_j$ .

Definition 4.7: Overlap segment. In G, consider an ordered set of edges  $E_{os} = \{e_1, \ldots, e_l\} \subset E$  that forms a path. This path is called an overlap segment for paths  $P_i$  and  $P_j$  if

- (i)  $\forall k \in \{1, ..., l\}$ ,  $e_k$  is an overlap edge for  $P_i$  and  $P_j$ .
- (ii) None of the incoming edges into  $tail(e_1)$  are overlap edges for  $P_i$  and  $P_j$ .
- (iii) None of the outgoing edges leaving head $(e_l)$  are overlap edges for  $P_i$  and  $P_j$ .

Our solution strategy is as follows. We first convert the original instance into another *structured* instance where each internal node has at most degree three (in-degree + out-degree). We then convert this new instance into a minimal one, and then develop the code assignment algorithm. It will be evident that using this network code, one can obtain a network code for the original instance.

1) Conversion procedure: Let G = (V, E) be our original graph, and let  $s_i$  and  $t_i$  be the given sources and terminals. We can efficiently construct a *structured* graph  $\hat{G} = (\hat{V}, \hat{E})$  in which each internal node  $v \in \hat{V}$  is of total degree at most

three with the additional following properties: (a)  $\hat{G}$  is acyclic. (b) For every source (terminal) in G there is a corresponding source (terminal) in  $\hat{G}$ . (c) For any two edge disjoint paths  $P_i$  and  $P_j$  for one unicast session in G, there exist two *vertex* disjoint paths in  $\hat{G}$  for the corresponding session in  $\hat{G}$ . (d) Any feasible network coding solution in  $\hat{G}$  can be efficiently turned into a feasible network coding solution in G. Our reduction steps are the same as in [12]. Due to space limitations, refer to [12] and [13] for details.

2) Code Assignment Procedure: In the discussion below, we will assume that the graph G is structured. It is clear that this is without loss of generality based on the previous arguments. In our arguments, we will use the minimality of the graph extensively.

Lemma 4.8: Consider a minimal multiple unicast instance,  $\langle G, \{s_1-t_1, s_2-t_2\} \rangle$  with connectivity level  $[1\ m]$ . Denote the  $s_1-t_1$  path by  $P_1$  and the set of edge disjoint  $s_2-t_2$  paths by  $\{P_{21},\ldots,P_{2m}\}$ . There can be at most one overlap segment between  $P_1$  and each  $P_{2i}, i=1,\ldots,m$ .

*Proof:* Suppose that there are two overlap segments  $E_{os1} = \{e_1, \ldots, e_{k_1}\}$  and  $E_{os2} = \{e'_1, \ldots, e'_{k_2}\}$  between  $P_1$  and  $P_{2i}$ , where  $e_{k_1}$  is upstream of  $e'_1$ . Note that by the definition of an overlap segment and the fact that G is structured, it holds that the head of  $e_{k_1}$  has in-degree one and out-degree two, so that one outgoing edge from  $head(e_{k_1})$  belongs to  $P_1$  (denoted  $e^*$ ) and the other belongs to  $P_{2i}$ . Note also  $e^* \in P_1$  cannot belong to  $P_{2j}$ ,  $j \neq i$  since the set of paths  $\{P_{21}, \ldots, P_{2m}\}$  is vertex disjoint (since G is structured).

Now, note that  $e^*$  can be removed while still maintaining the required connectivity level. This is true for  $s_2-t_2$ , since  $e^*$  does not lie on any of the paths  $P_{21},\ldots,P_{2m}$ . It is true for  $s_1-t_1$  since there is a path from  $e_{k_1}$  to  $e'_{k_2}$  that overlaps  $P_{2i}$ , and therefore this still continues be a path from  $s_1-t_1$ . This path can be explicitly specified as  $path(s_1,head(e_{k_1})), path(e_{k_1},e'_{k_2}), path(head(e'_{k_2}),t_1)$ . Using this property, we can obtain the following result that holds for the case of two unicast sessions with the rate  $\{1,m\}$ .

Lemma 4.9: A minimal multiple unicast instance  $< G, \{s_1-t_1, s_2-t_2\}, \{1, m\} >$  with connectivity level [1 m+1] is always feasible.

*Proof:* We show that this can be achieved by using scalar linear network codes. Let  $P_1$  denote the path from  $s_1-t_1$  and m+1 vertex-disjoint paths from  $s_2-t_2$ , as  $P_{2j}, j=1,\ldots,m+1$ . Let the source message at  $s_1$  be denoted by  $X_1$  and the source message vector at  $s_2$  by  $[X_{21},\ldots,X_{2m}]$ . We proceed by induction on m.

Base case - m = 1. In this case suppose that  $P_1$  intersects at most one path from the  $s_2 - t_2$ . For instance if  $P_1$  overlaps with  $P_{21}$ , then simply transmit  $X_{21}$  over  $P_{22}$  and  $X_1$  over  $P_1$ .

Alternatively,  $P_1$  overlaps both  $P_{21}$  and  $P_{22}$ . Suppose that the segments are denoted  $E_{os1}$  and  $E_{os2}$  respectively and that  $E_{os1}$  is upstream of  $E_{os2}$  (w.l.o.g.). In this case, we flow  $X_1$  ( $X_{21}$ ) on  $P_1$  ( $P_{21}$ ) until  $E_{os1}$  and flow  $X_1 + X_{21}$  on  $E_{os1}$ , and further downstream on  $P_{21}$  till  $t_2$  and on  $P_1$  until  $E_{os2}$ . We flow  $X_{21}$  on  $P_{22}$  until  $E_{os2}$  and flow  $X_1 + X_{21} + X_{21} = X_1$ ,

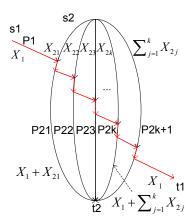


Fig. 4. An example where  $P_1$  overlaps with all paths  $P_{21},\ldots,P_{2k+1}$ . Rate  $\{1,k\}$  is feasible.

on  $E_{os2}$  and further downstream till  $t_1$  and  $t_2$ . It is evident that  $t_2$  can recover  $X_{21}$  from its received values.

Induction step. Suppose that the induction hypothesis holds for m=k. For m=k+1, again we consider two cases. Suppose that  $P_1$  does not overlap with at least one path from the set  $\{P_{21}, \ldots, P_{2k+1}\}$ , w.l.o.g. suppose that it is  $P_{2k+1}$ . In this case the graph consisting of  $P_1 \cup P_{21} \cup \cdots \cup P_{2k}$ , can be used to transmit  $X_1$  to  $t_1$  and  $X_{21}, \ldots, X_{2k-1}$  to  $t_2$  using the induction hypothesis.  $X_{2k}$  can simply be routed on  $P_{2k+1}$ .

On the other hand if  $P_1$  overlaps with all the paths  $P_{21},\ldots,P_{2k+1}$ . We assume w.l.o.g. that it overlaps first with  $P_{21}$  (in  $E_{os1}$ ), then with  $P_{22}$  and so on until  $P_{2k+1}$ . In this case, as illustrated in Figure 4, we can arrive at the required solution. In particular,  $s_2$  transmits  $X_{2i}$  over paths  $P_{2i}, i=1,\ldots,k$  and  $\sum_{j=1}^k X_{2j}$  over  $P_{2k+1}$  until the overlap point. The path  $P_1$  carries  $X_1$  until  $E_{os1}$ . At each overlap segment a sum of the incoming values into the segment is computed. This ensures that overlap segment  $E_{osi}$  carries  $X_1 + \sum_{j=1}^i X_{2j}, i=1,\ldots,k$  and  $E_{osk+1}$  carries  $X_1$ . It can be seen that both  $t_1$  and  $t_2$  are satisfied in this case.

It turns out that one can treat the case of three multiple unicast sessions with connectivity level [1 3 3], by using the result of Lemma 4.9. The basic idea is to use vector linear network coding over two time units and code over pairs of sources at appropriately defined layers of the network. We state and prove this result below.

Theorem 4.10: A multiple unicast instance with three sessions such that the connectivity level is [1 3 3] is always feasible.

*Proof:* Let the original graph (with unit capacity edges) be denoted by G=(V,E). We use vector linear network coding over two time units, i.e. T=2. In this case we form a new graph  $G^*$  where each edge  $e \in E$  is replaced by two parallel unit capacity edges  $e^1$  and  $e^2$  in  $G^*$ . The messages at source node  $s_i$  are denoted  $[X_{i1} \ X_{i2}]$ . Now consider the subgraph of  $G^*$  induced by all edges with superscript 1, that we denote  $G_1^*$ . In  $G_1^*$ , there exists a single  $s_1 - t_1$  path and three edge disjoint  $s_2 - t_2$  paths. Therefore, we can transmit

 $X_{11}$  from  $s_1$  to  $t_1$  and  $[X_{21} \ X_{22}]$  from  $s_2 - t_2$  using the result of Lemma 4.9. Similarly, we use the subgraph of  $G^*$  induced by all edges with superscript 2, i.e.,  $G_2^*$  to communicate  $X_{12}$  from  $s_1$  to  $t_1$  and  $[X_{31} \ X_{32}]$  from  $s_3$  to  $t_3$ . Thus, using vector linear network coding over two time units, a connectivity level of  $[1\ 3\ 3]$  suffices to satisfy the demands of each terminal.

Corollary 4.11: A multiple unicast instance with three sessions such that the connectivity level is greater than [1 3 3] is always feasible.

*Proof:* For the graph G which has connectivity level greater than  $[1\ 3\ 3]$ , we identify a subgraph G' with connectivity level  $[1\ 3\ 3]$ . By Theorem 4.10, the demand at each terminal can be satisfied. Then by assigning zero coding vector to the edges in  $G\setminus G'$ , the terminal demand can be satisfied in the original graph G.

So far, we have completely characterized the cases where the connectivity levels are  $[k_1 \ k_2 \ k_3]$ ,  $k_i \leq 3$ . However, there are several connectivity levels with unknown feasibility when  $k_i > 3$ , e.g.,  $[2\ 2\ 4]$ .

# V. CONCLUSIONS AND FUTURE WORK

In this work, we have identified several feasible and infeasible connectivity levels for 3 unicast sessions. For the feasible instances, we provided explicit network code assignments, while for the infeasible instances we demonstrated appropriate counter-examples. Some of these results can be extended to the case of general n, and are currently under investigation.

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